## CSE 332

JUNE 26TH - ANALYSIS OF THE HEAP

## ASSORTED MINUTIAE

- Some problems with EX02 and EX03


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- Checkpoint moved to Friday


## TODAY'S LECTURE

- bigO and analysis


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- Analyzing the heap


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- bigO and analysis
- Analyzing the heap
- Floyd's method


## REVIEW FROM LAST WEEK

- Heap implementation


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- Heap implementation
- Complete


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- Complete
- Heap property


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- Array implementation (0 or 1 indexing)


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- d-heaps


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- Algorithm analysis


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- bigO - asymptotic runtime bounds


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- Algorithm analysis
- Counting operations strictly is unreliable
- Want some way for us to compare functions
- bigO - asymptotic runtime bounds
- $f(n)=O(g(n))$ if there exists some $c$ and $n_{0}$ such that $f(n)<c * g(n)$ for some $\mathrm{c}>0$ and all $\mathrm{n}>\mathrm{n}_{0}$


## BIG-O NOTATION

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Formally, a function $f(n)$ is $\Omega(g(n))$ if there exists acc and $n_{0}>0$ such that:

- For all $n \geq n_{0}, f(n)>c * g(n)$
- If a function $f(n)$ is in $O(g(n))$ and今( $\mathrm{g}(\mathrm{n})$ )


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- If a function $f(n)$ is in $\mathbf{O ( g ( n ) )}$ and $\Omega(g(n))$, then $g(n)$ is a tight bound on $f(n)$, we call this big theta.
- Formally, if $f(n)$ is in $O(g(n))$ and $\Omega(g(n))$, then $f(n)$ is in $\theta(g(n))$
- Note that the two will have different c and $n_{0}$


## BIG O NOTATION

- What does this help us with?
- Sort algorithms into families


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- What does this help us with?
- Sort algorithms into families
- $O(1)$ : constant
- $\mathrm{O}(\log \mathrm{n})$ : logarithmic
- $O(n)$ : linear
- $O\left(n^{2}\right)$ : quadratic
- $O\left(n^{k}\right)$ : polynomial
- $\mathrm{O}\left(\mathrm{k}^{\mathrm{n}}\right)$ : exponential


## BIG $O$ NOTATION

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- What does this help us with?
- The constant multiple c lets us organize similar algorithms together.
- Remember that $\log _{\mathrm{a}} \mathrm{k}$ and $\log _{\mathrm{b}} \mathrm{k}$ differ by a constant factor?


## BIG O NOTATION

- What does this help us with?
- The constant multiple c lets us organize similar algorithms together.
- Remember that $\log _{\mathrm{a}} \mathrm{k}$ and $\log _{\mathrm{b}} \mathrm{k}$ differ by a constant factor?
- That makes all logs in the same family


## CORRECTNESS ANALYSIS

- How do we show an algorithm is correct?


## CORRECTNESS ANALYSIS

- How do we show an algorithm is correct?
- Need to look at the approach


## BINARY SEARCH (AGAIN)

```
public int binarySearch(int[] data, int toFind){
```

int low = 0; int high = data.length-1;
while(low <= high) \{
int mid = (low+high)/2;
if(toFind>mid) low = mid+1; continue;
else if(toFind<mid) high = mid-1; continue;
else return mid;
\}
return -1;
\}

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- Prove binary search returns the correct answer


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- If $A[i] \leq A[j]$, then $i \leq j$


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- Binary search always chooses the correct side


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- Prove binary search returns the correct answer
- Need property of sortedness
- For all pairs $\mathrm{i}, \mathrm{j}$ in the array:
- If $A[i] \leq A[j]$, then $i \leq j$
- Binary search always chooses the correct side
- End case: low = high


## ANALYSIS

- Let's use these analytical approaches to solve some things about heap functions


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- Let's use these analytical approaches to solve some things about heap functions
- First, let's do a quick review of heap properties


## REVIEW

- Is this a heap?



## REVIEW

- Is this a heap?
- No. Why?


REVIEW

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## REVIEW

- Is this a heap?



## REVIEW

- Is this a heap?
- Yes, Heap + Complete



## REVIEW

- Heaps
- Properties
- Completeness
- Heap property
- Implementation
- Array (0 v 1 indexing)


## REVIEW

- Array property



## REVIEW

- Array property



## REVIEW

- Array property


|  | $\mathbf{A}$ | $\mathbf{B}$ | $\mathbf{C}$ | $\mathbf{D}$ | $\mathbf{E}$ | $\mathbf{F}$ | $\mathbf{G}$ | $\mathbf{H}$ | $\mathbf{I}$ | $\mathbf{J}$ | $\mathbf{K}$ | $\mathbf{L}$ |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 | 13 |

## REVIEW

- With 0 indexing:
- Parent:
- Left-child:
- Right-child:


## REVIEW

- With 0 indexing:
- Parent: (i-1)/2
- Left-child:
- Right-child:


## REVIEW

- With 0 indexing:
- Parent: ( $\mathrm{i}-1$ )/2
- Left-child: $2 \mathrm{i}+1$
- Right-child:


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## REVIEW

- With 1 indexing:
- Parent:
- Left-child:
- Right-child:


## REVIEW

- With 1 indexing:
- Parent: i/2
- Left-child:
- Right-child:


## REVIEW

- With 1 indexing:
- Parent: i/2
- Left-child: 2 i
- Right-child:


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- With 1 indexing:
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- Left-child: 2 i
- Right-child: $2 \mathbf{i + 1}$


## REVIEW

- What about for a d-heap?


## REVIEW

- What about for a d-heap?
- Arithmetic changes slightly, but it is still doable


## HEAPS

- Operations
- Insert: adds a data, priority pair into the heap


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- Operations
- Insert: adds a data, priority pair into the heap
- deleteMin: returns and removes the item of smallest priority from the heap
- changePriority: changes the priority of a particular item in the heap
- What are the (worst-case) runtimes for these operations?


## HEAPS

- Insert:
- Add the element at the bottom of the tree


## HEAPS

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- What is the height of a heap?


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- What is the height of a heap? $\log _{2} \mathbf{n}$


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- Add the element at the bottom of the tree
- "Percolate up" that element to its correct place
- Adding to the end of a tree? $O(1)$
- Percolating up? O(height) $O(\log n)$
- What is the height of a heap? $\log _{2} \mathbf{n}$


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## HEAPS

- changePriority:


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- changePriority:
- Find the element
- Percolate up/down


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- Finding in a heap? O(n) Why?
- Heap property does not give us the divide and conquer benefit
- Percolate up/down? O(log $n$ )
- On average, is it faster to percolate up or down?


## ANALYSIS

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- Given an array of length $\mathbf{n}$, how do we make that array into a heap?
- Naïve approach?
- Make a new heap and add each element of the array into the heap
- How long to finish?


## FUN FACTS!

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- Early insertions are into empty trees $\mathbf{O}(1)$ !
- Consider a simpler example, creating a sorted linked list.
- Adding at the end of a linked list with $k$ items takes $\mathrm{O}(\mathrm{k})$ operations.
$1+2+3+4+5 \ldots$
What is this summation?


## FUN FACTS!

$$
\sum_{k=1}^{n} k=\frac{1}{2} n(n+1)
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- What does this mean?


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- What does this mean?
- Summing $k$ from 1 to $n$ is still $0\left(n^{2}\right)$


## FUN FACTS!

$$
\sum_{k=1}^{n} k=\frac{1}{2} n(n+1)
$$

- What does this mean?
- Summing $k$ from 1 to $n$ is still $0\left(n^{2}\right)$
- Similarly, summing log(k) from 1 to $n$ is O(n $\log n$ )


## ANALYSIS

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- Must add n items


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- Each add takes how long?


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- Naïve approach:
- Must add n items
- Each add takes how long? log(n)
- Whole operation is $O(\log (n))$
- Can we do better?
- What is better? $O(n)$


## HEAPS

- Facts of binary trees


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- Facts of binary trees
- Increasing the height by one doubles the number of possible nodes
- Therefore, a complete binary tree has half of its nodes in the leaves
- A new piece of data is much more likely to have to percolate down to the bottom than be the smallest item in the heap


## BUILDHEAP

- So a naïve buildheap takes $O(n \log n)$


## BUILDHEAP

- So a naïve buildheap takes $O(n \log n)$
- Why implement at all?


## BUILDHEAP

- So a naïve buildheap takes $O(n \log n)$
- Why implement at all?
- If we can get it $O(n)$ !


## FLOYD'S METHOD

- Traverse the tree from bottom to top
- Reverse order in the array


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- Start with the last node that has children.
- How to find?


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## FLOYD'S METHOD

- Traverse the tree from bottom to top
- Reverse order in the array
- Start with the last node that has children.
- How to find? Size / 2
- Percolate down each node as necessary
- Wait! Percolate down is $O(\log n)$ !
- This is an $\mathrm{O}(\mathrm{n} \log \mathrm{n})$ approach!


## FLOYD'S METHOD

- It is $O(n \log n)$, because big $O$ is an upper bound, but there is a tighter analysis possible!


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## FLOYD'S METHOD

- It is $O(n \log n)$, because big $O$ is an upper bound, but there is a tighter analysis possible!
- How far does each node travel (at worst)
- Leaves don't move at all: Height $=0$
- This is half the nodes in the tree


## FLOYD'S METHOD

- It is $O(n \log n)$, because big $O$ is an upper bound, but there is a tighter analysis possible!
- How far does each node travel (at worst)
- $1 / 2$ of the nodes don't move:
- These are leaves - Height $=0$
- 1/4 can move at most one
- 1/8 can move at most two


## FLOYD'S METHOD

- It is $O(n \log n)$, because big $O$ is an upper bound, but there is a tighter analysis possible!
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## FLOYD'S METHOD

$$
\sum_{i=0}^{n} \frac{i}{2^{i+1}}=2^{-n-1}\left(-n+2^{n+1}-2\right)
$$

- Thanks Wolfram Alpha!


## FLOYD'S METHOD

$$
\sum_{i=0}^{n} \frac{i}{2^{i+1}}=2^{-n-1}\left(-n+2^{n+1}-2\right)
$$

- Thanks Wolfram Alpha!
- Does this look like an easier summation?


## FLOYD'S METHOD

$$
\sum_{i=0}^{\infty} \frac{1}{2^{i+1}}=1
$$

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- This is a must know summation!


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- This is a must know summation!
- $1 / 2+1 / 4+1 / 8+\ldots=1$


## FLOYD'S METHOD

$$
\sum_{i=0}^{\infty} \frac{1}{2^{i+1}}=1
$$

- This is a must know summation!
- $1 / 2+1 / 4+1 / 8+\ldots=1$
- How do we use this to prove our complicated summation?


## FLOYD'S METHOD

$$
1 / 2+1 / 4+1 / 8 \ldots \quad \ldots+1 / 2^{n}=1
$$

## FLOYD'S METHOD

$$
\begin{array}{rll}
1 / 2+1 / 4+1 / 8 \ldots & \ldots+1 / 2^{n} & =1 \\
1 / 4+1 / 8 \ldots & \ldots+1 / 2^{n} & =1 / 2 \\
1 / 8 \ldots & \ldots+1 / 2^{n} & =1 / 4
\end{array}
$$

## FLOYD'S METHOD

$1 / 2+1 / 4+1 / 8$... ... $+1 / 2^{n}=1$

$$
\begin{array}{rll}
1 / 4+1 / 8 \ldots & \ldots+1 / 2^{n}=1 / 2 \\
1 / 8 \ldots & \ldots+1 / 2^{n}=1 / 4
\end{array}
$$

- Vertical columns sum to:
$i / 2^{\wedge} i$, which is what we want
- What is the right summation?
- Our original summation plus 1


## FLOYD'S METHOD

$$
\sum_{i=1}^{\infty} \frac{i}{2^{i}}=2
$$

## FLOYD'S METHOD

$$
\sum_{i=1}^{\infty} \frac{i}{2^{i}}=2
$$

- This means that the number of swaps we perform in Floyd's method is 2 times the size... So Floyd's method is $0(n)$


## NEXT LECTURE

- Back to analysis


## NEXT LECTURE

- Back to analysis
- Recurrences and analyzing recursive functions

